

UNIVERSITY OF CALIFORNIA
Los Angeles

Closed-Loop Subspace Identification of a Quadrotor

A thesis submitted in partial satisfaction
of the requirements for the degree
Master of Science in Engineering

by

Andrew G. Kee

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ABSTRACT OF THE THESIS

Closed-Loop Subspace Identification of a Quadrotor

by

Andrew G. Kee

Master of Science in Engineering

University of California, Los Angeles, 2013

Professor Steve Gibson, Chair

Ne quo feugiat tractatos temporibus, eam te malorum sensibus. Impetus voluptua
senserit id mel, lucilius adipiscing at duo. Suas noster sanctus cu pro, movet
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The thesis of Andrew G. Kee is approved.

Steve Gibson, Committee Chair

University of California, Los Angeles

2013

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NOMENCLATURE

N	Number of block rows
s	Number of block columns
\mathbb{R}	Set of all real numbers
u	System input
y	System output
x	System state

CHAPTER 1

Introduction

Unmanned Aerial Vehicles (UAVs) have seen explosive growth in the past thirty years, performing a multitude of military and civilian tasks including surveillance, reconnaissance, armed combat operations, search and rescue, forest fire management, and domestic policing [9, 10]. A class of modern UAVs which have recently grown in popularity are quadrotors - Vertical Take Off and Landing (VTOL) vehicles powered by four rotors arranged in a cross configuration. The main advantage of the quadrotor lies in its mechanical simplicity. Adjusting the speed of one or more of the vehicle's fixed-pitch rotors provides full attitude control, eliminating the need for the swash plate mechanism found on single rotor helicopters [1, 2]. In spite of its mechanical simplicity, the quadrotor exhibits somewhat complex dynamics that are best modeled as a Multi-Input Multi-Output (MIMO) system.

Advances in MEMS sensors and light-weight high-powered lithium polymer batteries have contributed to the recent popularity of quadrotors, making them an attractive choice for research applications in flight dynamics and control, as in [3, 5, 6, 7]. One problem of particular interest is the development of mathematical models representing system dynamics based on experimentally gathered data. System identification provides a mechanism to relate this input-output data to the underlying system dynamics. Traditionally, system identification techniques have focused on developing a system model which minimizes prediction error. Identification methods of this form are commonly known as Prediction Error Methods (PEMs). PEMs have seen widespread use in both theoretical and real-world ap-

plications, but experience difficulties with MIMO systems as noted in [8, 11]. Subspace identification methods have recently grown in popularity and offer an alternative approach to the identification problem. These methods have a foundation in linear algebra and overcome the issues found in PEMs when identifying MIMO systems [4]. It is the goal of this research project to apply subspace identification techniques to a quadrotor using experimentally gathered closed-loop input and output data.

1.1 Related Work

For text, let's use the first words out of the ispell dictionary.

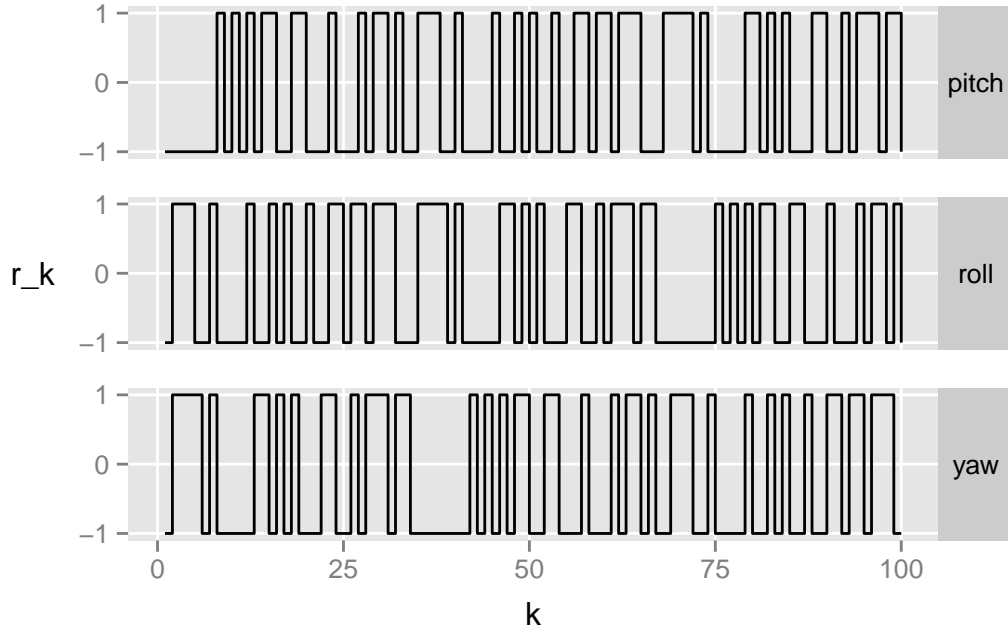


Figure 1.1: A figure caption

1.2 Motivation and Contributions

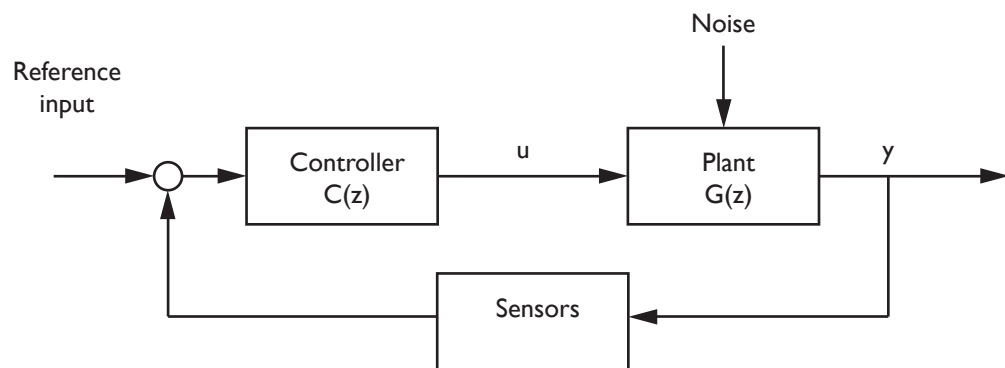


Figure 1.2: A figure caption

CHAPTER 2

Model and Assumptions

Introduce concept of LTI system here

2.1 State Space Model

We will consider a combined deterministic-stochastic LTI system written in innovation form as

$$x_{k+1} = Ax_k + Bu_k + Ke_k \quad (2.1a)$$

$$y_k = Cx_k + Du_k + e_k \quad (2.1b)$$

where $x_k \in \mathbb{R}^n$ is the system state, $u_k \in \mathbb{R}^m$ is the system input, $y_k \in \mathbb{R}^l$ is the system output, and $e_k \in \mathbb{R}^l$ is the innovation. A , B , C , and D are the system matrices with appropriate dimensions and K is the Kalman filter gain. The system represented in (2.1) can also be represented in predictor form as

$$x_{k+1} = A_K x_k + B_K u_k + K y_k \quad (2.2a)$$

$$y_k = Cx_k + Du_k + e_k \quad (2.2b)$$

where $A_K = A - KC$ and $B_K = B - KD$.

The systems represented by (2.1) and (2.2) are equivalent from an input/output point of view, but because A_K is guaranteed stable even if the original process matrix A is unstable, the predictor form proves advantageous when considering unstable open-loop systems.

2.2 Assumptions

[**Assumption 1**]: $A_K = A - KC$ is stable (i.e. its eigenvalues lie within the unit circle)

[**Assumption 2**]: The system is represented in its minimal form

CHAPTER 3

Subspace Identification Methods

3.1 Subspace Identification

3.1.1 General Subspace Identification Problem

3.1.2 Geometric Tools

Matrices with Hankel structure Matrix Row/Column space Projections

3.1.3 Subspace Identification of Combined Deterministic-Stochastic Systems

3.2 Closed-Loop Subspace Identification

3.2.1 Identifying Systems Operating Under Feedback Control

3.2.2 Innovation Estimation Method

3.2.3 Whitening Filter

CHAPTER 4

Experiments

4.1 Experiment Design

CHAPTER 5

Results

CHAPTER 6

Conclusion

6.1 Future Work

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